# Notes of Probability and Stochastics

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## Chapter 1

# Measure and integrations

## 1.1 Measurable space

## 1.1.1 $\sigma$ algebra

**Definition 1.1.** A nonempty system of subset of  $\Omega$  is an algebra on  $\Omega$  if it's

- 1. Closed under complement:  $A^c \in \mathcal{A} \iff A \in \mathcal{A}$
- 2. Closed under finite union:  $\cup_i A_i \in \mathcal{A} \iff \forall i, A_i \in \mathcal{A}$

it's an  $\sigma$  algebra on  $\Omega$  if it's also closed under countable union.

Remark.  $\mathcal{A}$  is an algebra auto implies  $\emptyset \in \mathcal{A}$  and  $\Omega \in \mathcal{A}$ . So  $\{\emptyset, \Omega\}$  is the minimum algebra on  $\Omega$  and thus minimum  $\sigma$  algebra while the discrete algebra  $2^{\Omega}$  is maximum.

Let  $\{\mathcal{A}_{\gamma} : \gamma \in \Gamma\}$  is a collection of  $\sigma$  algebra, then we have  $\mathcal{A} = \cap_{\gamma \in \Gamma} \mathcal{A}_{\gamma}$  is also a  $\sigma$  algebra. Hence we can define the smallest  $\sigma$  algebra as intersection of all  $\sigma$  algebras contains  $\mathcal{A}$ , that called the  $\sigma$  algebra **generated** by  $\mathcal{A}$  and denoted by  $\sigma(\mathcal{A})$ .

The smallest  $\sigma$ -algebra generated by the system of all open sets in a topological space  $(\Omega, \tau)$  is called **Borel**  $\sigma$  **algebra** on  $\Omega$  and denoted by  $\mathcal{B}(\Omega)$ , its elements are called **Borel sets**.

### 1.1.2 $\pi, \lambda, m$ systems

A collection of subsets  $\mathcal{A}$  is called.

- 1. **m-system** if closed under monotone series, that is if  $(A_n) \subset \mathcal{A}$  and  $A_n \nearrow A$ , then  $A \in \mathcal{A}$ .
- 2.  $\pi$ -system is closed under finite intersection
- 3.  $\lambda$  system if
  - 1.  $\Omega \in \mathcal{A}$
  - 2. closed under proper difference:

$$A - B \in \mathcal{A} \iff B \subset A \text{ and } B, A \in \mathcal{A}$$

3. is an m-system.

**Theorem 1.1.** Let  $\mathcal{A}$  be a collection of subsets of  $\Omega$  iff it's both a  $\pi$  system and  $\lambda$  system.

Which can be proved as follows:

- **⇒**:
  - 1.  $\Omega \in \mathcal{A}$
  - $2. \ A-B=A\cap B^c\in \mathcal{A}$
  - 3. is an m-system
- =:
  - 1.  $A^c = \Omega A \in \mathcal{A}$
  - 2.  $A \cup B = (A^c \cap B^c)^c \in \mathcal{A}$
  - 3. hence  $\mathcal{A}$  is an algebra and  $\mathcal{A}$  is a m-system.

Similarly, for  $m, \pi, \lambda$  -system, they also has a minimum system generated by some collection  $\mathcal{C}.$ 

**Lemma 1.1.** Let  $\mathcal{A}$  be an algebra, then

- 1.  $m(A) = \sigma(A)$
- 2. if  $\mathcal{B}$  is an m class and  $\mathcal{A} \subset \mathcal{B}$ , then  $\sigma(\mathcal{A}) = m(\mathcal{A}) \subset \mathcal{B}$

Similarly, let  $\mathcal{A}$  be a  $\pi$  class, then  $\lambda(\mathcal{A}) = \sigma(\mathcal{A})$ 

Then we have **Monotone class theorem**:

**Theorem 1.2.**  $\forall \mathcal{A} \subset \mathcal{B} \subset \mathcal{P}(\Omega), s.t.$ :

- 1. If  $\mathcal{A}$  is a  $\pi$ -class,  $\mathcal{B}$  is a  $\lambda$ -class, then  $\sigma(\mathcal{A}) \subset \mathcal{B}$
- 2. If  $\mathcal{A}$  is an algebra,  $\mathcal{B}$  is a m-class, then  $\sigma(\mathcal{A}) \subset \mathcal{B}$

## 1.1.3 Measurable spaces

**Definition 1.2** (Measurable Space). Pair  $(\Omega, \mathcal{A})$  where  $\mathcal{A}$  is a  $\sigma$ -Algebra on  $\Omega$ .

**Definition 1.3** (Products of measurable spaces). Let  $(E, \mathcal{E}), (F, \mathcal{F})$  be two measurable spaces. For  $A \subset E, B \subset F, A \times B$  is the set of all pairs  $(x, y) : x \in A, y \in B$ . Note that  $\mathcal{E} \times \mathcal{F}$  is also a  $\sigma$ -Algebra with all  $A \times B$  where  $A \in \mathcal{E}, B \in \mathcal{F}$  which is called the product  $\sigma$ -Algebra.

## 1.2 Measurable function

## 1.2.1 Mappings

Let  $X:\Omega_1\to\Omega_2$  be a mapping,  $\forall B\subset\Omega_2$  and  $\mathcal{G}\subset\mathcal{P}(\Omega_2)$ , the **inverse image** of

• 
$$B \text{ is } X^{-1}(B) = \{\omega : \omega \in \Omega_1, X(\omega) \in B\} := \{X \in B\}$$

$$\bullet \ \mathcal{G} \text{ is } X^{-1}(\mathcal{G}) = \left\{X^{-1}(B) : B \in \mathcal{G}\right\}$$

There is some properties:

1. 
$$X^{-1}(\Omega_2) = \Omega_1, X^{-1}(\varnothing) = \varnothing$$

2. 
$$X^{-1}(B^c) = [X^{-1}(B)]^c$$

3.

$$\begin{array}{l} X^{-1}\left(\cup_{\gamma\in\Gamma}B_{\gamma}\right)=\cup_{\gamma\in\Gamma}X^{-1}\left(B_{\gamma}\right) \text{ for } B_{\gamma}\subset\Omega_{2}, \gamma\in\Gamma\\ X^{-1}\left(\cap_{\gamma\in\Gamma}B_{\gamma}\right)=\cap_{\gamma\in\Gamma}X^{-1}\left(B_{\gamma}\right) \text{ for } B_{\gamma}\subset\Omega_{2}, \gamma\in\Gamma \end{array}$$

where  $\Gamma$  may not countable.

4. 
$$X^{-1}(B_1 - B_2) = X^{-1}(B_1) - X^{-1}(B_2) \forall B_1, B_2 \subset \Omega_2$$

5. 
$$B_1 \subset B_2 \subset \Omega_2 \implies X^{-1}(B_1) \subset X^{-1}(B_2)$$

- 6. If  $\mathcal{B}$  is a  $\sigma$ -algebra,  $X^{-1}(\mathcal{B})$  is also a  $\sigma$ -algebra. It's easy to check  $X^{-1}(\mathcal{B})$  is closed under complement and countable union. (From properties 2 and 3)
- 7. If  $\mathcal{C}$  is nonempty,  $X^{-1}(\sigma(\mathcal{C})) = \sigma(X^{-1}(\mathcal{C}))$

**Remarks**  $X^{-1}$  preserves all the set operations on  $\Omega$ .

#### 1.2.2 Measurable functions

**Definition 1.4.** For two measurable spaces  $(\Omega_1, \mathcal{A})$ ,  $(\Omega_1, \mathcal{B})$ ,  $f : \Omega_1 \to \Omega_2$  is a **measurable mapping** if  $f^{-1}(\mathcal{B}) \subset \mathcal{A}$ , where

$$f^{-1}(\mathcal{B})=\{f^{-1}(B):B\in\mathcal{B}\}$$

It is a measurable function if  $(\Omega_2, \mathcal{B}) = (\mathcal{R}^n, \mathcal{B}(\mathcal{R}^n))$ , moreover, a Borel function if  $(\Omega_1, \mathcal{A}) = (\mathcal{R}^m, \mathcal{B}(\mathcal{R}^m))$ 

Remark. If  $\mathcal{B} = \sigma(\mathcal{C})$ , the definition can be reduced to  $X^{-1}(\mathcal{C}) \subset \mathcal{A}$  since

$$X^{-1}(\mathcal{B}) = X^{-1}(\sigma(\mathcal{C})) = \sigma\left(X^{-1}(\mathcal{C})\right) \subset \sigma(\mathcal{A}) = \mathcal{A}$$

**Lemma 1.2.** Suppose  $f: \mathcal{E} \to \mathcal{F}$  and  $g: \mathcal{F} \to \mathcal{G}$  are measurable, then so is  $f \circ g$ .

 ${\it Proof.}$  The same as how we proved composition of continuous function is continuous.

## 1.3 Random Variable

A r.v. X is a measurable function from  $(\Omega_1, \mathcal{A})$  to  $(\mathcal{R}, \mathcal{B})$ . It denoted by X is  $\mathcal{A}$ -measurable or  $X \in \mathcal{A}$ 

(Another definition): A r.v. X is a measurable mapping from  $(\Omega, \mathcal{A}, P)$  to  $(\mathcal{R}, \mathcal{B})$  such that

$$P(|X| = \infty) = P(\{\omega : |X(\omega)| = \infty\}) = 0$$

**Lemma 1.3.** X is a r.v. from  $(\Omega, \mathcal{A})$  to  $(\mathbb{R}, \mathcal{B})$ 

$$\iff X \leq x = X^{-1}([-\infty], x) \in \mathcal{A} \quad \forall x \in \mathbb{R} \iff X \leq x = X^{-1}([-\infty], x) \in \mathcal{A} \quad \forall x \in D$$

where D is a dense subset of  $\mathbb{R}$ , e.g.  $\mathbb{Q}$ .  $\{X \leq x\}$  above can be replaced by

$$\{X \le x\}, \quad \{X \ge x\}, \quad \{X < x\}, \quad \{X > x\}, \quad \{x < X < y\}$$

#### 1.3.1 Construction of random variables

**Lemma 1.4.**  $\mathbf{X} = (X_1, \dots, X_n)$  is a random vectors if  $X_k$  is a r.v.  $\forall k$  iff  $\mathbf{X}$  is a measurable function from  $(\Omega, \mathcal{A})$  to  $(\mathcal{R}^n, \mathcal{B}(\mathcal{R}^n))$ .

Proof. Note that

$$\{\mathbf{X}\in\prod I_n\}=\bigcap\{X_n\in I_n\}\in\mathcal{A}$$

where  $I_k=(a_k,b_k], -\infty \leq a_k \leq b_k \leq \infty$  and follows from  $\sigma(\{\prod I_n\})=\mathcal{B}(\mathcal{R}^n)$ . For the other direction, note

$$\{X_k \leq t\} = \{\mathbf{X} \in \prod_{i < k} \mathbb{R} \times \{-\infty, t\} \times \prod_{i > k} \mathbb{R}\} \in \mathcal{A}. \blacksquare$$

Recall lemma 1.2 we have:

**Theorem 1.3.**  $\forall$  random n vectors  $X = (X_{1:n})$  and Borel function f from  $\mathcal{R}^n \to \mathcal{R}^m$ , then f(X) is a random m vectors.

Remark. Note that continuous function are borel measurable since continuity leads to inverse image of an open set is still open. So if  $X_{1:n}$  are r.v.'s, so are  $\sum X_n$ ,  $\sin(x)$ ,  $e^X$ ,  $\operatorname{Poly}(X)$ , .... That implies:

$$\forall X,Y \in \mathcal{A}, \text{ so are } aX + bY, X \lor Y = \max\{X,Y\}, X \land Y = \min\{X,Y\}, X^2, XY, X/Y, X^+ = \max(x,0), X^- = -\min(x,0), |X| = X^+ + X^-$$

## 1.3.2 Limiting opts

Let  $(X_n)$  are r.v. on  $(\Omega, \mathcal{A})$ , then  $\sup_{n \to \infty} X_n$ ,  $\inf_{n \to \infty} X_n$ ,  $\lim\sup_{n \to \infty} X_n$ ,  $\liminf_{n \to \infty} X_n$  are r.v.'s. Moreover, if it exists,  $\lim_{n \to \infty} X_n$  is r.v..

*Proof.* First two follows from,  $\forall t \in \mathbb{R}$ :

$$\{\sup_{n\to\infty}X_n\leq t\}=\bigcap_{n=1}^\infty\{X_n\leq t\}\in\mathcal{A}$$

$$\{\inf_{n\to\infty}X_n\geq t\}=\bigcap_{n=1}^\infty\{X_n\geq t\}\in\mathcal{A}$$

and the last two follows from  $\limsup_{n\to\inf}=\inf_{k\to\infty}\sup_{m\geq k}X_m$  and  $\liminf_{n\to\inf}=\sup_{k\to\infty}\inf_{m\geq k}X_m.$ 

That implies

**Lemma 1.5.** If  $S = \sum_{1}^{\infty} X_n$  exists everywhere, then S is a r.v.

Proof. Note  $\sum_{1}^{\infty} X = \lim_{n \to \infty} \sum_{n} X_n$  is a r.v.

If  $X = \lim_{n \to \infty} X_n$  holds **almost** everywhere, i.e., define  $\Omega_0$  is the set of all  $\omega$ , such that  $\lim_n X_n(\omega)$  exists, then  $P(\Omega_0) = 1$ , we say that  $X_n$  converges a.s. and write:

$$X_n \to X$$
 a.s.

For a measurable function f, we may modify it at a null set into f' and it remain mesurable since for any open set G,  $f'^{-1}(G)$  differ  $f^{-1}(G)$  a at most null set, by the completion of lebesgue measure space, f'-1(G) is measurable and thus  $f^{-1}$  measurable. Hence, for  $f_n \to f$  a.s., we may ignore a null set and then  $f_n \to f$  everywhere and thus f measurable.

## 1.3.3 Approximations of r.v. by simple r.v.'s

**Definition 1.5.** If  $A \in \mathcal{A}$ , the indicator function  $\mathbf{1}_A$  is a r.v.

If  $\Omega = \sum_{1}^{n} A_i$ , where  $A_i \in \mathcal{A}$ , then  $X = \sum_{1}^{n} a_i I_{A_i}$  is a r.v. and called **simple** r.v.

Any r.v. can be approximated by simple ones:

**Theorem 1.4.**  $\forall X \in \mathcal{A}, \ \exists 0 \leq X_1 \leq X_2 \cdots X_n \ s.t. \ X_n(\omega) \nearrow X(\omega) \ everywhere.$ 

Proof. Suppose

$$X_n(\omega) = \sup\{\frac{j}{2^n}: j \in \mathbb{Z}, \frac{j}{2^n} \leq \min(X(\omega), 2^n)\}$$

One can check  $X_n$  is simple r.v. and  $X_n(\omega)\nearrow X(\omega)$  for all  $\omega\in\Omega.$ 

## 1.3.4 $\sigma$ algebra generated by r.v.

Let  $\{X_{\lambda}, \lambda \in \Lambda\}$  is r.v.s on  $(\Omega, \mathcal{A})$ . Define

$$\sigma\left(X_{\lambda},\lambda\in\Lambda\right):=\sigma\left(X_{\lambda}\in B,B\in\mathcal{B},\lambda\in\Lambda\right)=\sigma\left(X_{\lambda}^{-1}(\mathcal{B}),\lambda\in\Lambda\right)=\sigma\left(\cup_{\lambda\in\Lambda}X_{\lambda}^{-1}(\mathcal{B})\right)$$

which is called  $\sigma$  algebra generated by  $\{X_{\lambda}, \lambda \in \Lambda\}$ , where  $\Lambda$  is a index set which can be uncountable.

For  $\Lambda = \mathbb{N}^+$ :

1. 
$$\sigma\left(X_{i}\right)=\sigma\left(X_{i}^{-1}(\mathcal{B})\right)=X_{i}^{-1}(\mathcal{B})=\left\{X_{i}\in\mathcal{B}\right\}$$
 
$$\sigma\left(X_{1},\ldots,X_{n}\right)=\sigma\left(\cup_{i=1}^{n}X_{i}^{-1}(\mathcal{B})\right)=\sigma\left(\cup_{i=1}^{n}\sigma\left(X_{i}\right)\right)$$

2. 
$$\begin{split} \sigma\left(X_{1}\right) \subset \sigma\left(X_{1}, X_{2}\right) \subset \ldots \ldots \subset \sigma\left(X_{1}, \ldots, X_{n}\right) \\ \sigma\left(X_{1}, X_{2}, \ldots \ldots\right) \supset \sigma\left(X_{2}, X_{3}, \ldots \ldots\right) \supset \ldots \ldots \supset \sigma\left(X_{n}, X_{n+1}, \ldots \ldots\right) \end{split}$$

3.  $\bigcap_{1}^{\infty} \sigma(X_n, X_{n+1}, \cdots)$  is the tail  $\sigma$  algebra of  $X_1$ .

If  $A_{1:n}$  are not mutually exclusive to each other, then we have

$$|\sigma(A_{1:n})| = 2^{2^n}$$

Which follows from for a partition  $A_{1:n}$ ,

$$\sigma(A_1,\cdots,A_n)=\{\bigcup_{i\in J}A_i\}$$

where J is any subset of  $\mathbb{N} \leq n$  and  $A_0 = \emptyset$ . Hence for discrete r.v. Y,  $\sigma(Y)$  can be generated from  $A_i = \{Y = y_i\}$  for all  $y_i$ 's. For continuous case, it's generated by all intervals.

### 1.3.5 Monotone classes of function

**Definition 1.6** (monotone class).  $\mathcal{M}$  is called a monotone class if:  $-1 \in \mathcal{M} - f, g \in \mathcal{M}_b$  and  $a, b \in \mathbb{R} \implies af + bg \in \mathcal{M} - (f_n) \subset M_+, f_n \uparrow f \implies f \in \mathcal{M}$ 

where  $\mathcal{M}_+$  is a subcollection consisting of positive functions in  $\mathcal{M}$ , and  $\mathcal{M}_b$  for the bounded function in  $\mathcal{M}$ .

**Theorem 1.5** (Monotone class theorem for functions). Let  $\mathcal{M}$  be a monotone class of functions on E. Suppose for some p-system  $\mathcal{C}$  generating  $\mathcal{E}$  and  $1_A \in \mathcal{M}$  for every  $A \in \mathcal{C}$ . Then  $\mathcal{M}$  includes all positive  $\mathcal{E}$ -measurable functions and all bounded  $\mathcal{E}$ -measurable functions.

*Proof.* First we need to show that  $1_A \in \mathcal{M}$  for every  $A \in \mathcal{E}$ . Let  $\mathcal{D} = \{A \in \mathcal{E} : 1_A \in \mathcal{M}\}$ . Now we check that  $\mathcal{D}$  is a d-system:  $-1_E = 1$ , so  $E \in \mathcal{D}$ .  $-B \subset A$ ,  $A, B \in \mathcal{D}$ .  $1_{A-B} = 1_A - 1_B \in \mathcal{D}$   $-(A_n) \subset \mathcal{D}$  and  $A_n \uparrow A$ , then  $1_{A_n} \uparrow 1_A$ , so  $1_A \in \mathcal{M}$ , then  $A \in \mathcal{D}$ 

By assuption,  $\mathcal{C} \subset \mathcal{D}$ , and  $\sigma(\mathcal{C})$  is the smallest d-system by the theorem above, so  $\mathcal{E} \subset \mathcal{D}$ , so  $1_A \in \mathcal{M}$  for every  $A \in \mathcal{E}$ .

As  $1_A \in \mathcal{M}$  for every  $A \in \mathcal{E}$ , we can easily prove that all of the positive simple function is generated by the linear combination of  $1_A$  s. And all positive  $\mathcal{E}$ -measurable functions is generated by a sequence of positive simple functions.

Then for general bounded  $\mathcal{E}$ -measurable function f, using  $f = f^+ - f^-$  where  $f^+, f^- \in \mathcal{M}$ .

Let  $(E,\mathcal{E}),(F,\mathcal{F})$  be two mesurable spaces and f is a bijection  $E\to F$ . Then f is said to be a isomorphism of  $(E,\mathcal{E})$  and  $(F,\mathcal{F})$  if f is  $\mathcal{E}$ -measurable and  $f^{-1}$  is  $\mathcal{F}$ -measurable. These two spaces are called isomorphic if there exists an isomophism between them.

A measurable apce  $(E,\mathcal{E})$  is said to be standard if it is isomorphic to  $(F,\mathcal{B}_F)$ for some Borel subset  $F \subset \mathbb{R}$ .

#### 1.4Measure

Let  $\Omega$  be a space and  $\mathcal{A}$  a class, then function  $\mu: \mathcal{A} \to R = [-\infty, \infty]$  is a set function.

It's

- 1. **finite** if  $\forall A \in \mathcal{A}$ ,  $|\mu(A)| < \infty$ 
  - 2.  $\sigma$ -finite if  $\exists A_n \subset \mathcal{A}, \quad s.t. \quad \cup_{i=1}^{\infty} A_i = \Omega \quad \forall n \quad |\mu(A_n)| < \infty$ 3.  $\Sigma$  finite if there exist finite  $(\mu_n)$  s.t.  $\mu = \sum_n \mu_n$ .
- 1. additive  $\iff \mu\left(\sum_{i=1}^{n}A_{i}\right) = \sum_{i=1}^{n}\mu\left(A_{i}\right)$ 2.  $\sigma$ -additive $\iff \mu\left(\sum_{i=1}^{\infty}A_{i}\right) = \sum_{i=1}^{\infty}\mu\left(A_{i}\right)$

*Remark.* Finite implies  $\sigma$  finite and  $\sigma$  finite implies  $\Sigma$  finite.

 $\mu$  is a **measure** on  $\mathcal{A}$  if

- 1.  $\forall A \in \mathcal{A} : \mu(A) \geq 0$
- 2. It's  $\sigma$  additive.

the triplet  $(\Omega, \mathcal{A}, \mu)$  is a **measure space** when  $\mu$  is a measure and  $(\Omega, \mathcal{A})$  is a measurable space. Whose sets are called **measurable sets** or  $\mathcal{A}$ -measurable.A measure space is a **probability space** if  $P(\Omega) = 1$ .

Assume that  $A_{1:n} \in \mathcal{A}$  and  $A \in \mathcal{A}$  and  $\mu$  is a measure.

- 1.  $\mu$  is continues from above, if  $A_n \searrow A \implies \mu(A_n) \rightarrow \mu(A)$ 2.  $\mu$  is continues from below, if  $A_n \nearrow A \implies \mu(A_n) \rightarrow \mu(A)$
- 3.  $\mu$  is continues at A, if  $A_n \to A \implies \mu(A_n) \to \mu(A)$

 $\forall$  Measure  $\mu$  is continues from below and may not continues from above. It will be continues from above if  $\exists m < \infty, \mu(A_m) < \infty$ . So finite measure  $\mu$  are always continues.

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#### Properties of measure 1.4.1

#### 1.4.1.1 Semialgebras

Let  $\mu$  be a nonnegative additive set function on a semialgebra  $\mathcal{A}$ .  $\forall A, B \in \mathcal{A}$ and  $\{A_n, B_n, n \ge 1\} \in \mathcal{A}$ 

- 1. (Monotonicity):  $A \subset B \implies \mu(A) \leq \mu(B)$
- 2.  $(\sigma$ -subadditivity):
  - $\begin{array}{ll} 1. \ \sum_{1}^{\infty}A_{n}\subset A, &\Longrightarrow \ \sum_{1}^{\infty}\mu\left(A_{n}\right)\leq\mu(A) \\ 2. \ \ \text{Moreover, if} \ \mu \ \text{is a measure, then} \end{array}$

$$B \subset \sum_{n=1}^{\infty} B_n \implies \mu(B) \leq \sum_{n=1}^{\infty} \mu\left(B_n\right)$$

We can assert a nonnegative set function  $\mu$  is a measure by:

- 1.  $\mu$  is additive
- 2.  $\mu$  is  $\sigma$  subadditive on  $\mathcal{S}$

### 1.4.1.2 Algebras

Let  $\mu$  be a measure on an algebra  $\mathcal{A}$ 

$$A \subset \cup_1^\infty A_n \implies \mu(A) \leq \sum_1^\infty \mu\left(A_n\right)$$

**Proof** Note  $A=A\cap (\cup A_n)=\cup (A\cap A_n),$  hence we can write A as union in  $\mathcal A$ by take  $B_n = A \cap A_n \in \mathcal{A}$ .

$$A = \bigcup_{1}^{\infty} B_n$$

and then we can take  $C_n = B_n - \cup_1^{n-1} B_i \in \mathcal{A}$  to write A as disjoint union:

$$A = \sum C_n$$

Then

$$\mu(\mathcal{A}) = \mu(\sum C_n) = \sum \mu(C_n) \leq \sum \mu(B_n) \leq \sum \mu(A_n)$$

as 
$$C_n \subset B_n \subset A_n$$
.

#### 1.4.1.3 $\sigma$ algebras

Let  $\mu$  be a measure on an  $\sigma$  algebra  $\mathcal{A}$ 

- 1. Monotonicity
- 2. Boole's inequality(Countable Sub-Additivity)

$$\mu\left(\cup_{i=1}^{\infty}A_{i}\right)\leq\sum_{i=1}^{\infty}\mu\left(A_{i}\right)$$

- 3. Continuity from below
- 4. Continuity from above if  $\mu$  is finite in  $A_i$ .

The sense of 4 follows from suppose  $A_i \searrow A$ , then  $A_1 - A_i \nearrow A_1 - A$ , then

$$\mu(A_1) - \mu(A) = \mu(A_1 - A) = \lim \mu(A_1 - A_i) = \mu(A_1) - \lim(A_i)$$

where  $\mu(A_1)$  cannot be cancelled if  $\mu(A_i) = \infty$ .

**Definition 1.7.** Let  $(\Omega, \mathcal{A}, \mu)$  be a measure space, and  $N \subset \Omega$ 

- 1. N is a  $\mu$  null set iff  $\exists B \in \mathcal{A}$  s.t.  $\mu(B) = 0$ ,  $N \subset B$
- 2. This measure space is a **complete measure** space if  $\forall$   $\mu$  null space N,  $N \in \mathcal{A}$

**Theorem 1.6.** Given any measure space  $(\Omega, \mathcal{A}, \mu)$ , there exist a complete measure space  $(\Omega, \bar{\mathcal{A}}, \bar{\mu})$ , such that  $\mathcal{A} \subset \bar{\mathcal{A}}$  and  $\bar{\mu}$  is an extension of  $\mu$ . This space is called completion of  $(\Omega, \mathcal{A}, \mu)$ .

Proof. Take

$$\begin{split} \bar{\mathcal{A}} &= \{A \cup N : A \in \mathcal{A}\} \\ \bar{\mathcal{B}} &= \{A \Delta N : A \in \mathcal{A}\} \end{split}$$

 $\bar{\mathcal{A}} = \bar{\mathcal{B}}$  since  $A \cup N = (A - B)\Delta(B \cap (A \cup N))$  and  $A\Delta N = (A - B)\cup(B \cap (A\Delta N))$ .

Then we can show that  $\bar{\mathcal{A}}$  is a  $\sigma$  algebra. Let  $E_i = A_i \cup N_i \in \bar{\mathcal{A}}$ , then

$$\bigcup_1^\infty E_i = \bigcup_1^\infty A_i \cup \bigcup_1^\infty N_i$$

and note  $\bigcup_1^{\infty} A_i \in \mathcal{A}$  and  $\mu(\bigcup_1^{\infty} N_i) \leq \mu(\bigcup_1^{\infty} B_i) \leq \bigcup_1^{\infty} \mu(B_i) = 0$ . Thus  $\bar{\mathcal{A}}$  is closed by countable union. As for complements, note  $E^c = A^c \cap N^c = (A^c \cap N^c \cap B^c) \cup (A^c \cap N^c \cap B) = (A^c \cap B^c) \cup (A^c \cap N^c \cap B) \in \bar{\mathcal{A}}$ .

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Finally we define a measure  $\bar{\mu}$  on  $\bar{\mathcal{A}}$  by

$$\bar{\mu}(A \cup N) = \mu(A)$$

We should prove it's well defined. Suppose  $A_1 \cup N_1 = A_2 \cup N_2 \in \bar{\mathcal{A}}$ , note  $A\Delta B\Delta C = A\Delta (B\Delta C)$  and  $A\Delta B = B\Delta A$ .

$$\begin{split} (A_1 \Delta A_2) \Delta (N_1 \Delta N_2) &= (A_1 \Delta A_2 \Delta N_1) \Delta N_2 \\ &= (A_1 \Delta N_1) \Delta (A_2 \Delta N_2) \\ &= \varnothing \end{split}$$

Hence  $A_1\Delta A_2=N_1\Delta N_2$ , note  $N_1\Delta N_2\subset N_1\cup N_2\subset B_1\cup B_2$ , hence  $\mu(A_1\Delta A_2)=0$  and thus  $\mu(A_1-A_2)=\mu(A_2-A_1)=0$ . Therefore

$$\begin{split} \mu(A_1) &= \mu(A_1 - A_2) + \mu(A_1 \cap A_2) = \mu(A_1 \cap A_2) \\ \mu(A_2) &= \mu(A_2 - A_1) + \mu(A_1 \cap A_2) = \mu(A_1 \cap A_2) \end{split}$$

 $\bar{\mu}$  is do well defined.  $\mu^*$  is auto  $\sigma$  additive since so is  $\mu$  and is easy to check that all  $\mu^*$  null set is  $\mu$  null set.

### 1.4.2 Specification of measures

**Theorem 1.7.** Let  $(E,\mathcal{E})$  be a measurable space. Let  $\mu,\nu$  be measures on it with  $\mu(E) = \nu(E) < \infty$ . If  $\mu,\nu$  agree on a p-system generating  $\mathcal{E}$ , then  $\mu,\nu$  are identical. >

Proof. Let  $\mathcal C$  be the p-system generating  $\mathcal E$  and  $\mu(A)=\nu(A)$  for every  $A\in\mathcal C$ . Consider  $\mathcal D=\{A\in\mathcal E:\mu(A)=\nu(A)\}$  which satisfies  $\mathcal C\subset\mathcal D\subset\mathcal E$ . Then we need to prove that  $\mathcal D$  is a d-system:  $-E\in\mathcal D$  by the assuption. - Let  $A,B\in\mathcal D$  and  $B\subset A$ . Then  $\mu(A-B)=\mu(A)-\mu(B)=\nu(A)-\nu(B)=\nu(A-B)$ , so  $A-B\in\mathcal D$  - Let  $(A_n)\uparrow A$  and  $(A_n)\subset\mathcal D$ , then  $\mu(A_n)\uparrow\mu(A),\nu(A_n)\uparrow\nu(A)$ , since  $\mu(A_n)=\nu(A_n)$  for every n, so  $\mu(A)=\nu(A)$ . So  $\mathcal D$  is a d-system.

So  $\mathcal{D} = \mathcal{E}$ , so for every  $A \in \mathcal{E}$ ,  $\mu(A) = \nu(A)$ .

Let  $(E, \mathcal{E})$  be a measurable space. Suppose that the sigleton  $\{x\} \in \mathcal{E}$  for every  $x \in E$ .

Let  $\mu$  be a measure on  $(E, \mathcal{E})$ . A point x is said to be an atom if  $\mu(\{x\}) > 0$ , the measure is said to be *diffuse* if it has no atoms. It is said to be *purely atomic* if the set D of its atoms is countable and  $\mu(E - D) = 0$ .

**Theorem 1.8.** Let  $\mu$  be a  $\Sigma$ -finite measure on  $(E, \mathcal{E})$ . Then  $\mu = \nu + \lambda$  where  $\lambda$  is a diffuse measure and  $\nu$  is purely atomic.